## A GENERALIZATION OF THE LAW OF LARGE NUMBERS

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It is well known that the law of large numbers can be established for dependent as well as for independent chance variables by using Tchebycheff's inequality [1] and assuming that the variance of the sum of the variables tends towards infinity less rapidly than  $n^2$ .

In recent years v. Mises has introduced the notion of statistical functions [2] and has shown that, under certain assumptions the law of large numbers is still valid if, instead of the arithmetic mean of the n observations  $x_1, \dots, x_n$  a statistical function of these observations is considered. For example in the very special case, where the n collectives which have been observed are identical k-valued arithmetic distributions with probabilities  $p_1, \dots, p_k$  corresponding to the attributes  $c_1, \dots, c_k$  and with observed relative frequencies  $n_1/n, \dots, n_k/n$  one obtains the result: It is to be expected for every  $\epsilon > 0$  with a probability  $P_n$  converging towards one as  $n \to \infty$ , that  $|f(n_1/n, \dots, n_k/n) - f(p_1, \dots, p_k)| < \epsilon$  under very general conditions concerning the function f.

In the present paper we shall generalize these new results so that they will apply also to collectives which are not independent.

1. Lemma concerning alternatives. Let us consider the *n*-dimensional collective consisting of a sequence of *n* trials and let us assume that the *n* trials are alternatives, i.e. for each trial there are only two possible results which we denote by "success," "failure," by "occurrence," "non-occurrence" or by "1," "0." The total result of the *n* trials is expressed by *n* numbers each equal to 0 or 1. Let  $v(x_1, x_2, \dots, x_n)$  be the probability of obtaining the result  $x_1$  at the first trial,  $x_2$  at the second one,  $\dots$ ,  $x_n$  at the last one  $(x_r = 0, 1; r = 1, \dots, n)$ . In the same way we introduce  $v_{12}(x, y) = \sum_{x_3, \dots, x_n} v(x, y, x_3, \dots, x_n)$  and generally  $v_{\mu r}(x, y)$  as the probability that the  $\mu$ th result equals x, the rth equals x, r and finally let r and finally let r be the probability that the r the probability that the r th result equals r. In particular let us write

$$v_{\mu}(1) = p_{\mu}, \quad v_{\mu\nu}(1, 1) = p_{\mu\nu}, \quad (\mu, \nu = 1, \dots, n; \mu \neq \nu)$$

 $p_{\mu}$  being the probability of success in the  $\mu$ th trial and  $p_{\mu\nu}$  the probability of simultaneous success both in the  $\mu$ th and  $\nu$ th trials.

The variance  $s_n^2$  of the sum  $(x_1 + \cdots + x_n)$  is easily found: